
Markov Modeling

Part 8

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Markov Models, Monte Carlo and Software Demonstration

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Using Excel to build a Markov Model

- Create a sheet for transition probabilities
- Put transition probability names across columns
- Put probabilities for subsequent years in subsequent rows

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Using Excel to build a Markov Model

- Create a sheet for health states
- Put names of health states across columns
- Populate “year 0” with desired starting populations
- Use transition probabilities to derive populations for each subsequent year

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Using Excel to build a Markov Model

- Sum “occupancy” of each health state
- Apply discounting formula if desired
- Multiply time spent in each state by cost for that state and utility for that state

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Using TreeAge

- Pros
 - Easy to build models
 - Good for model with many states
- Cons
 - Workings are a “black box” - difficult to understand the mechanics
 - More challenging to present to decision-makers

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Monte Carlo simulation

- Primary use: evaluate uncertainty
- A numerical integration technique
 - (originally devised by Feynman during the Manhattan Project)
- Can be time-consuming, but
- Gives most reasonable estimate of overall uncertainty in the model

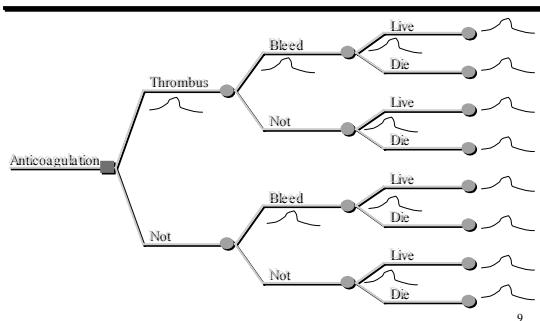
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Limitations of deterministic sensitivity analysis

- No standard for permissible level of variation
- Analyst controls
 - which parameters varied
 - and over what range
- Complex interactions not captured by 1-, 2-, or 3-way sensitivity analysis
- Extreme case scenarios not realistic

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MC simulation



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Probabilistic methods

- All costs, probabilities, and outcomes are associated with uncertainty simultaneously
- Certain values are “more likely”
- How do we calculate overall uncertainty in results?

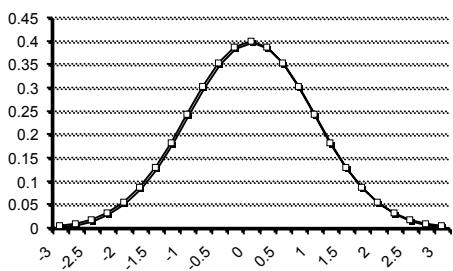
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Monte Carlo simulation

- Solve multiple integrals using numerical integration
- E.g.,
 - draw values randomly from all distributions
 - calculate expected values for tree
 - repeat until results converged

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Probability density



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Assigning distributions

- Probabilities and utilities

- must be bounded by 0 and 1
- use uniform, beta or logistic normal

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Logistic normal

- Logistic normal easy to use

- use formula of Doubilet et al
- provide mean and upper or lower bound
- formula provides mean and s.d. of normal distribution that when plugged into logistic function, $\text{EXP}(y)/(1+\text{EXP}(y))$, produces distribution with desired properties

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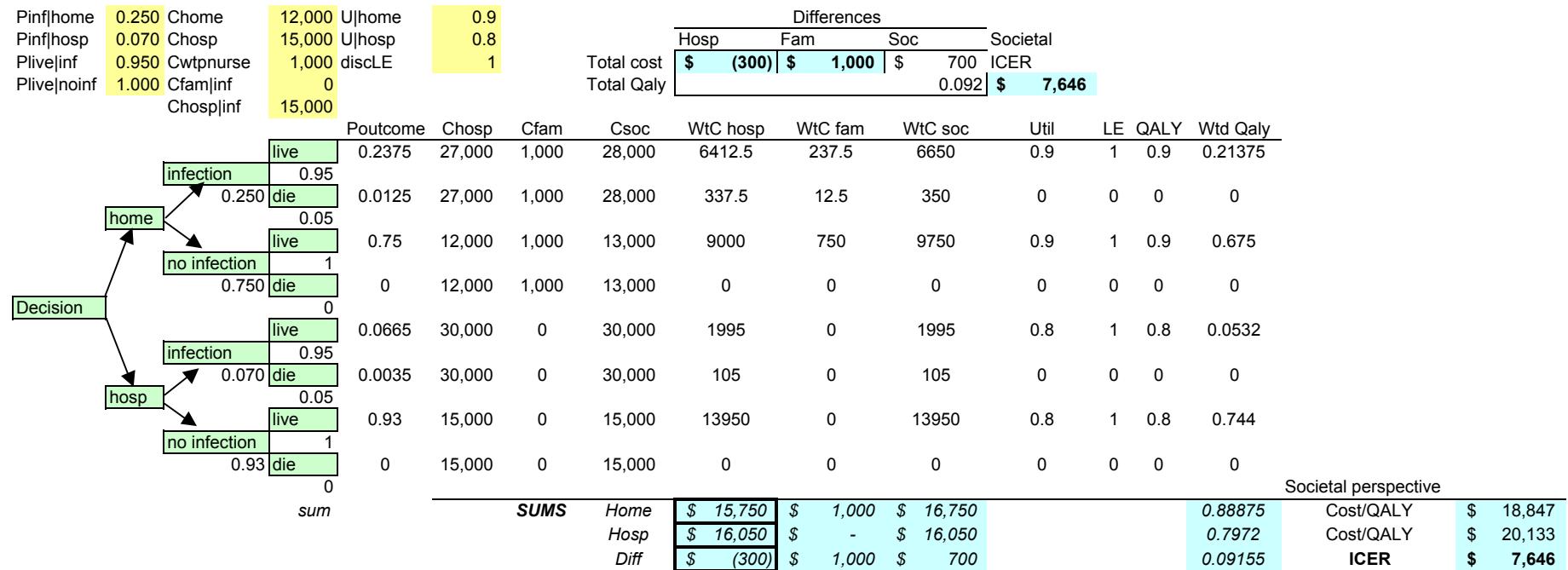
Assigning distributions

- Costs:

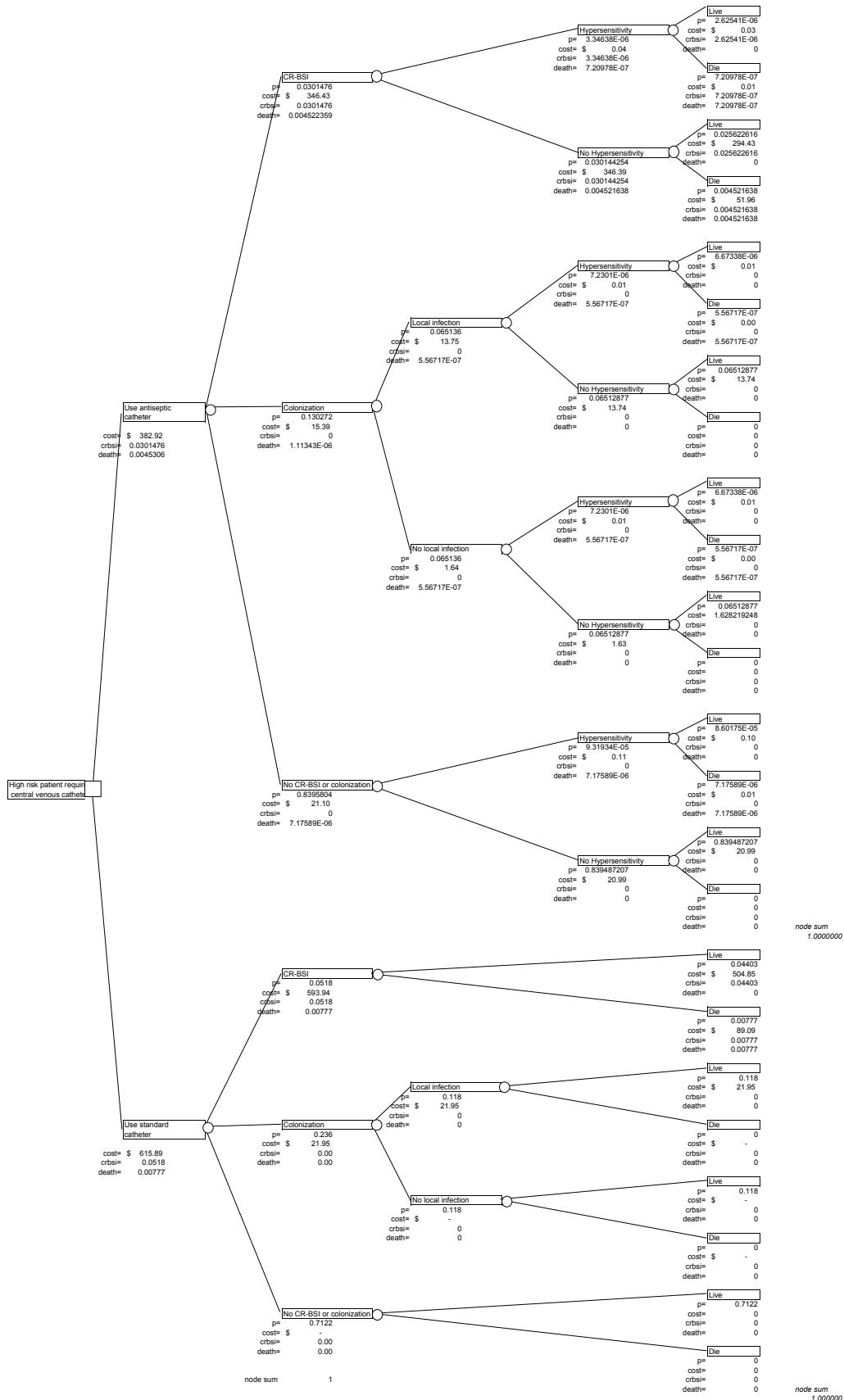
- distribution should reflect uncertainty in the population mean, not the sample mean - so is often normal even though (sample) costs are often skewed
- if range of possible population means is skewed, consider gamma function

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Decision Tree



Catheter Decision Tree



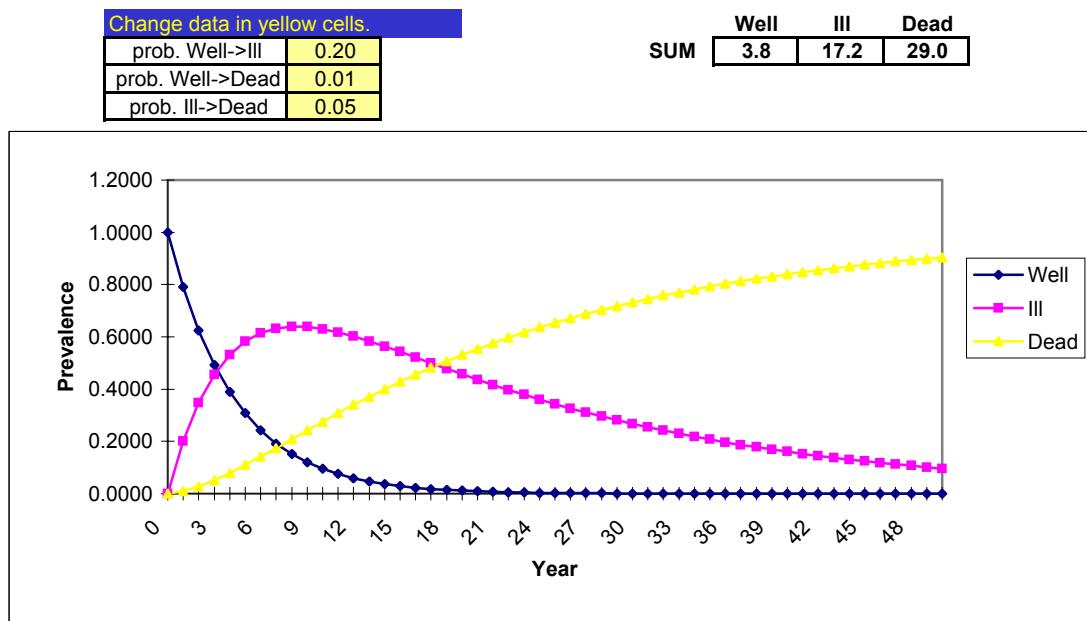
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node sum 1.000000

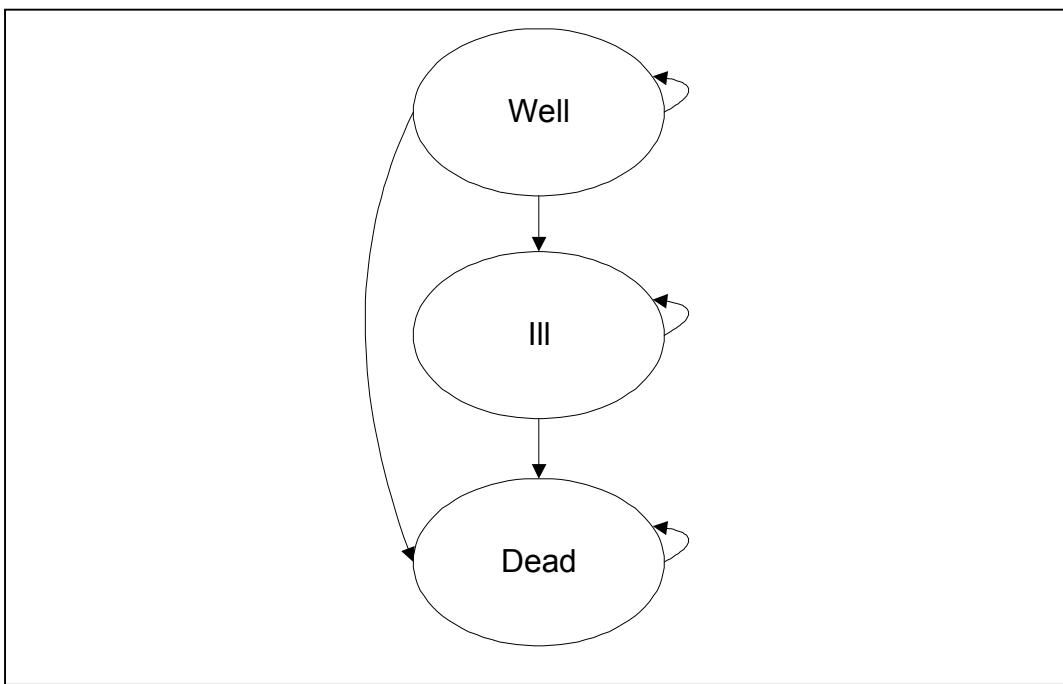
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3-state Markov Model

Example of a 3-state Markov Model



Model Diagram

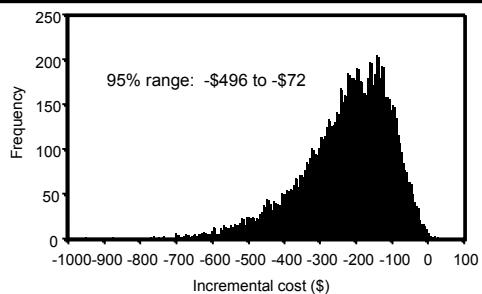


Evaluating results

- Once simulation is complete, look at distribution of *expected* results
- Determine 95% central range of results, or percentiles within quadrants on CE plane

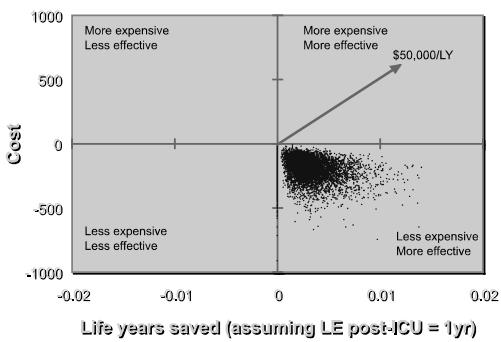
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Incremental cost



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Catheter simulation



Levels of uncertainty

- 1st order
 - sampling distribution
 - track outcomes of individual patients
 - no distributions used
- 2nd order
 - population distribution
 - track expected values for tree (population)
 - employ distributions

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Software

- TreeAge
- AtRisk simulation package in conjunction with Excel spreadsheet

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